

Foxulate - Word Mining Approach using BERT

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Abstract: In this paper, we propose a novel approach for automated keyword extraction using a combination of DistilBERT masking method and KeyBERT. We begin by using the DistilBERT model which is trained on a large corpus of text, using a masking strategy to identify the most informative tokens in each document. We then use the KeyBERT technique to create a list of keywords and key phrases that are most similar to the masked tokens in each document. Our approach is both minimal and easy-to-use, as it requires only a single model and does not rely on any additional external resources or heuristics. We evaluate our method on several benchmark datasets and demonstrate that it achieves state-of-the-art performance on a range of keyword extraction tasks. Our results show that our approach is both effective and efficient, and has the potential to be a valuable tool for a wide range of NLP applications.

Keywords: Deep Learning, DistilBERT, KeyBERT, Contextual Embeddings, Masking Method, Keyword Generation, Text Mining, Machine Learning, NLP Applications, Language Models, Unsupervised Learning

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